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Topological sensitivity and FMM-accelerated BEM applied to 3D acoustic inverse scattering

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Abstract

This study is set in the framework of inverse scattering of scalar (e.g. acoustic) waves. A qualitative probing technique based on the distribution of topological sensitivity of the cost functional associated with the inverse problem with respect to the nucleation of an infinitesimally-small hard obstacle is formulated. The sensitivity distribution is expressed as a bilinear formula involving the free field and an adjoint field associated with the cost function. These fields are computed by means of a boundary element formulation accelerated by the Fast Multipole method. A computationally fast approach for performing a global preliminary search based on the available overspecified boundary data is thus defined. Its usefulness is demonstrated through results of numerical experiments on the qualitative identification of hard obstacles in a bounded 3-D acoustic domain, for configurations featuring $O(10^5)$ nodal unknowns and $O(10^6)$ sampling points, based on exact or noisy synthetic data.

Key words: topological derivative, Helmholtz equation, inverse scattering, boundary elements, fast multipole method

1 Introduction

Defect identification problems are often solved by minimization of a cost function featuring the experimental data and (if available) prior information. Such

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cost functions are non-convex and exhibit local minima. Despite that fact, traditional iterative minimization or equation-solving methods are usually preferred to global search techniques such as evolutionary algorithms due to the prohibitive computational cost of solving large numbers of forward wave scattering problems. To perform optimally, gradient-based iterative algorithms are used in conjunction with shape sensitivity techniques see e.g. [22,26,28].

Still, the stand-alone use of gradient-type minimization for such purposes is not satisfactory for its success is strongly dependent on a reliable prior information about the geometry of the hidden object. This has prompted the development of so-called ‘sampling’ or ‘probe’, non-iterative, methods. They may be used either in isolation or as a preliminary step for choosing adequate initial guesses to be used in subsequent standard optimization schemes. Such methods are surveyed in a recent review article [29] and include the linear sampling method [7,25], not pursued here, and the application of topological sensitivity, which is the subject of this article. The concept of topological sensitivity consists in quantifying the sensitivity of a given cost functional with respect to the nucleation of an infinitesimal obstacle at (many) specified locations in the reference (background) medium. The topological sensitivity therefore allows to define a spatial obstacle indicator. This concept initially appeared first in [11,30] in connection with topological optimization of mechanical structures, allowing to define algorithms where “excess” material is iteratively removed until a satisfactory shape and topology is reached [14]. More recently, other investigations have studied the topological sensitivity as a sampling tool for inverse scattering problems, in the context of identification of cavities in 3D semi-infinite and infinite elastic media [18] and in elastic 3D bounded bodies [5], of penetrable inclusions in acoustic media [6] or elastic bodies [20], and for problems formulated in the time domain [19] (see also [13] for 2D elastostatics and [12] for 2D linear acoustics).

The derivation of topological sensitivity formulae for a given cost functional requires an asymptotic expansion of the scattered field induced by the nucleating obstacle. Although related asymptotic expressions have been proposed elsewhere (e.g. in [1] for inverse electromagnetic problems and penetrable scatterers, or in the framework of the low-frequency limit in scattering theory [10]), they have so far been utilized for the *explicit* reconstruction of *small* inhomogeneities based on either i) treatment of the measurement residuals [2,34,35], or ii) the reciprocity-gap approach [3]. In contrast, the present approach, that utilizes the spatial distribution of topological sensitivity as an indicator function, aims at *approximate* reconstruction of finite obstacles and may be considered as a sampling technique exploiting asymptotic expansions.

The distribution of topological sensitivity can be expressed in terms of a bilinear formula involving the free field and an adjoint field associated with the cost function. However, the computational cost of solving the forward and ad-

joint problems and evaluating the topological sensitivity distribution on a fine sampling grid increases rapidly with the non-dimensional wavenumber. The purpose of this article is to propose the topological sensitivity field computed by means of the Fast Multipole BEM (FM-BEM) [16,17,27] as the basis of a computationally fast tool for probing acoustic media for hidden hard obstacles on the basis of overdetermined boundary data, within the model framework of forward scattering problems governed by the scalar Helmholtz equation. To that end, the FM-BEM is in particular applied to evaluate in a fast way the integral representation formulae expressing the free and adjoint fields at a large number of sampling points inside the medium.

This article is organized as follows. After some preliminaries concerning the forward and inverse problems of interest (Section 2), the concept of topological sensitivity is presented in Section 3. The FM-BEM approach for the scalar Helmholtz equation is then summarized in Section 4. Finally, results of numerical experiments on qualitative scatterer identification using computed distributions of topological sensitivity are presented in Section 5, for configurations featuring $O(10^5)$ nodal BE unknowns and $O(10^6)$ sampling points.

2 Forward and inverse problems

This article is concerned with the identification of rigid obstacles embedded in acoustic media. The generic acoustic scattering problem of interest is defined as follows. Let Ω denote a three-dimensional open domain, either bounded or unbounded, with a sufficiently regular boundary S and filled with an acoustic medium characterized by wave velocity c and mass density ρ ; this configuration will be referred to as the reference (i.e. obstacle-free) medium. Let B^* denote a rigid scatterer (or a set thereof) bounded by the closed surface Γ^* , so that $\Omega^* = \Omega \setminus \bar{B}^*$ is the acoustic region surrounding the scatterer. Steady-state excitations on S with angular frequency ω generate an acoustic pressure field u^* in the acoustic domain Ω^* , governed by the following set of field equations and boundary conditions (collectively denoted by $\mathcal{P}(B^*)$ for later reference):

$$\mathcal{P}(B^*) : \quad \begin{aligned} (\Delta + k^2)u^* &= 0 & (\text{in } \Omega^*), \\ p[u^*] &= p^D & (\text{on } S), \\ p[u^*] &= 0 & (\text{on } \Gamma^*), \end{aligned} \quad (1)$$

where $k = \omega/c$ is the wavenumber, $w \rightarrow p[w] \equiv w_{,n} = \nabla w \cdot \mathbf{n}$ denotes the normal derivative operator, \mathbf{n} is the normal on $S \cup \Gamma^*$ outward to Ω^* , and p^D is the prescribed Neumann data over S (other types of boundary conditions may be considered as well). The implicit time-harmonic factor $e^{-i\omega t}$ is, as usual, omitted for brevity. It is assumed that ω is not an eigenfrequency of any of the boundary-value problems arising in the ensuing developments.

In the inverse scattering problem of interest, an unknown obstacle B^{true} , of boundary Γ^{true} , is to be identified. The corresponding exact acoustic field u^{true} is then governed by problem $\mathcal{P}(B^{\text{true}})$. With reference to problem $\mathcal{P}(B^{\text{true}})$, supplementary information is needed for the identification of B^{true} . Here, measured values u^{obs} of acoustic pressure are assumed to be available over the measurement surface $S^{\text{obs}} \subset S$. Ideally (i.e. assuming that the physics is exactly described by the chosen linear acoustics setting and that no measurement errors are present), u^{obs} is the trace of u^{true} on S^{obs} . The identification of B^{true} may then be formulated in terms of the minimization of a cost function. Generic cost function of format

$$\mathcal{J}(\Omega^*) = \int_{S^{\text{obs}}} \varphi(u_{\text{R}}^*(\boldsymbol{\xi}), u_{\text{I}}^*(\boldsymbol{\xi}), \boldsymbol{\xi}) \, d\Gamma \quad (2)$$

are considered, where u^* is the boundary trace of the solution to the forward problem $\mathcal{P}(B^*)$ for an assumed obstacle configuration B^* , the subscripts 'R' and 'I' being used to indicate the real and imaginary parts of a complex quantity (i.e. $w_{\text{R}} = \text{Re}(w)$ and $w_{\text{I}} = \text{Im}(w)$). For instance, the output least-squares cost function associated to measurement u^{obs} on S^{obs} , commonly used for such purposes, corresponds to

$$\varphi(w_{\text{R}}, w_{\text{I}}, \boldsymbol{\xi}) = \frac{1}{2} |w(\boldsymbol{\xi}) - u^{\text{obs}}(\boldsymbol{\xi})|^2, \quad (3)$$

The minimization of such cost functions can be performed using many methods, all of which are iterative and need repeated evaluations of $J(\Omega^*)$. Traditional gradient-based minimization may converge within a moderate number of evaluations of $\mathcal{J}(\Omega^*)$ if the trial surface Γ^* can be described in terms of a few geometrical parameters, but reach a local minimum which depends on the choice of initial guess. Global search techniques, e.g. evolutionary algorithms [24] or sampling methods based on the Metropolis algorithm [33], perform a global search (i.e. identify absolute and/or multiple minima), but at the cost of very large numbers of cost functions evaluations. In this article, the topological sensitivity is proposed as a tool for performing a qualitative global search at a computational cost which is far below that entailed by a true global optimization technique.

3 Topological sensitivity of the cost function

3.1 Notations

Let $B_\varepsilon(\mathbf{x}_s) = \mathbf{x}_s + \varepsilon\mathcal{B}$, where $\mathcal{B} \subset \mathbb{R}^3$ is a fixed bounded open set with boundary \mathcal{S} and volume $|\mathcal{B}|$ containing the origin, define the region of space

occupied by a hard obstacle of (small) size $\varepsilon > 0$ containing a fixed sampling point \mathbf{x}_s . It is convenient to introduce the scaled position vector $\bar{\boldsymbol{\xi}}$ defined by

$$\boldsymbol{\xi} = \mathbf{x}_s + \varepsilon \bar{\boldsymbol{\xi}} \quad (\boldsymbol{\xi} \in B_\varepsilon, \bar{\boldsymbol{\xi}} \in \mathcal{B}) \quad (4)$$

In particular, this mapping recasts integrals over B_ε and Γ_ε into integrals over \mathcal{B} and \mathcal{S} , respectively, and transforms the differential volume and area elements according to

$$dV_\xi = \varepsilon^3 d\bar{V}_{\bar{\xi}} \quad (\boldsymbol{\xi} \in B_\varepsilon, \bar{\boldsymbol{\xi}} \in \mathcal{B}), \quad d\Gamma_\xi = \varepsilon^2 d\bar{\Gamma}_{\bar{\xi}} \quad (\boldsymbol{\xi} \in \Gamma_\varepsilon, \bar{\boldsymbol{\xi}} \in \mathcal{S}) \quad (5)$$

Without loss of generality, \mathbf{x}_s can be chosen as the center of B_ε , i.e. such that

$$\int_{B_\varepsilon} (\boldsymbol{\xi} - \mathbf{x}_s) dV_\xi = \mathbf{0}, \quad \text{i.e.} \quad \int_{\mathcal{B}} \bar{\boldsymbol{\xi}} d\bar{V}_{\bar{\xi}} = \mathbf{0}. \quad (6)$$

Let $u^\varepsilon = u^\varepsilon(\boldsymbol{\xi}; \mathbf{x}_s)$ denote the solution to the scattering problem $\mathcal{P}(B_\varepsilon(\mathbf{x}_s))$ defined by (1), where $\Omega^\star = \Omega_\varepsilon(\mathbf{x}_s) = \Omega \setminus \overline{B_\varepsilon(\mathbf{x}_s)}$ and $\overline{B_\varepsilon(\mathbf{x}_s)}$ is the closure of $B_\varepsilon(\mathbf{x}_s)$. Further, let $J(\varepsilon; \mathbf{x}_s)$ be defined by

$$J(\varepsilon; \mathbf{x}_s) = \mathcal{J}(\Omega_\varepsilon(\mathbf{x}_s)) = \int_{S^{\text{obs}}} \varphi(u_R^\varepsilon(\boldsymbol{\xi}; \mathbf{x}_s), u_I^\varepsilon(\boldsymbol{\xi}; \mathbf{x}_s), \boldsymbol{\xi}) d\Gamma, \quad (7)$$

For convenience, explicit references to \mathbf{x}_s will often be omitted in the sequel, e.g. by writing $J(\varepsilon)$ or $u^\varepsilon(\boldsymbol{\xi})$ instead of $J(\varepsilon; \mathbf{x}_s)$ or $u^\varepsilon(\boldsymbol{\xi}; \mathbf{x}_s)$.

The evaluation of $J(\varepsilon)$ entails solving for u^ε the forward problem $\mathcal{P}(B_\varepsilon)$. It is convenient, and customary, to decompose u^ε according to

$$u^\varepsilon = u + v^\varepsilon, \quad (8)$$

where u , the *free field* defined as the response of the obstacle-free (reference) medium Ω due to the given excitation p^D , solves

$$\begin{aligned} (\Delta + k^2)u &= 0 & (\text{in } \Omega), \\ p[u] &= p^D & (\text{on } S), \end{aligned} \quad (9)$$

while v^ε , the *scattered field*, solves

$$\begin{aligned} (\Delta + k^2)v^\varepsilon &= 0 & (\text{in } \Omega_\varepsilon), \\ p[v^\varepsilon] &= 0 & (\text{on } S), \\ p[v^\varepsilon] &= -p[u] & (\text{on } \Gamma_\varepsilon). \end{aligned} \quad (10)$$

3.2 Expansion of $J(\varepsilon)$

To establish the topological sensitivity of $J(\varepsilon)$, one starts with the expansion

$$J(\varepsilon) = J(0) + \int_{S^{\text{obs}}} \text{Re} \left[\varphi_{,u} v^\varepsilon \right] d\Gamma + o(|v^\varepsilon|_{S^{\text{obs}}}), \quad (11)$$

with

$$\varphi_{,u} \equiv \left(\frac{\partial \varphi}{\partial u_{\text{R}}^\varepsilon} - i \frac{\partial \varphi}{\partial u_{\text{I}}^\varepsilon} \right) \Big|_{u^\varepsilon = u}. \quad (12)$$

Let the *adjoint field* \hat{u} be defined by

$$\begin{aligned} (\Delta + k^2)\hat{u} &= 0 & (\text{in } \Omega), \\ p[\hat{u}] &= \varphi_{,u} & (\text{on } S^{\text{obs}}), \\ p[\hat{u}] &= 0 & (\text{on } S \setminus S^{\text{obs}}). \end{aligned} \quad (13)$$

Then, the reciprocity identity (i.e. third Green's formula) applied to the states \hat{u} and v^ε over the domain Ω_ε leads, by virtue of the boundary conditions in (10) and (13), to the identity

$$\int_{S^{\text{obs}}} \varphi_{,u} v^\varepsilon d\Gamma + \int_{\Gamma_\varepsilon} p[\hat{u}] v^\varepsilon d\Gamma + \int_{\Gamma_\varepsilon} \hat{u} p[u] d\Gamma = 0 \quad (14)$$

As a result, the integral in the r.h.s. of (11) is converted into integrals over the vanishing cavity surface. Besides, since both u and \hat{u} are also defined inside B_ε , the last integral in (14) can be recast into a domain integral over B_ε by means of the divergence formula. Expansion (11) then takes the form

$$J(\varepsilon) = J(0) + \text{Re} \left\{ \int_{B_\varepsilon} [\nabla u \cdot \nabla \hat{u} - k^2 u \hat{u}] dV - \int_{\Gamma_\varepsilon} v^\varepsilon p[\hat{u}] d\Gamma \right\} + o(|v^\varepsilon|_{S^{\text{obs}}}), \quad (15)$$

The first integral in (15) features a density function whose definition does not depend on ε , and its expansion about $\varepsilon = 0$ can therefore be obtained by simply using the scaled coordinates (4), (5) and expanding $[\nabla u \cdot \nabla \hat{u} - k^2 u \hat{u}](\mathbf{x}_s + \varepsilon \vec{\xi})$ about $\varepsilon = 0$. In contrast, the second integral of (15) features the scattered field v^ε , which depends on ε . Its asymptotic behaviour must then be obtained from that of v^ε on Γ_ε (taking into account the fact that Γ_ε also depends on ε). This step is based on exploiting an integral equation reformulation of equations (10).

3.3 Governing integral equation formulation for the scattered field

The governing problem (10) for the scattered field $v^\varepsilon = v^\varepsilon(\cdot; \mathbf{x}_s)$ can be recast as the boundary integral equation [4,8]:

$$\frac{1}{2}v^\varepsilon(\mathbf{x}) + \int_{\Gamma_\varepsilon} \mathcal{H}(\mathbf{x}, \boldsymbol{\xi}; k) v^\varepsilon(\boldsymbol{\xi}) \, d\Gamma_\xi = - \int_{\Gamma_\varepsilon} \mathcal{G}(\mathbf{x}, \boldsymbol{\xi}; k) p[u](\boldsymbol{\xi}) \, d\Gamma_\xi \quad (\mathbf{x} \in \Gamma_\varepsilon), \quad (16)$$

where the Green's function $\mathcal{G}(\mathbf{x}, \boldsymbol{\xi}, k)$ is defined by

$$(\Delta_\xi + k^2)\mathcal{G}(\mathbf{x}, \boldsymbol{\xi}, k) + \delta(\boldsymbol{\xi} - \mathbf{x}) = 0 \quad (\boldsymbol{\xi} \in \Omega), \quad \mathcal{H}(\mathbf{x}, \boldsymbol{\xi}, k) = 0 \quad (\boldsymbol{\xi} \in S), \quad (17)$$

and $\mathcal{H}(\mathbf{x}, \boldsymbol{\xi}; k) = \boldsymbol{\nabla} \mathcal{G}(\mathbf{x}, \boldsymbol{\xi}, k) \cdot \mathbf{n}(\boldsymbol{\xi})$ is the normal derivative of $\mathcal{G}(\mathbf{x}, \boldsymbol{\xi}, k)$ (the nabla symbol $\boldsymbol{\nabla}$, when used in front of such kernel, conventionally indicates a gradient with respect to the second argument $\boldsymbol{\xi}$). Moreover, the free and adjoint fields have the explicit expressions

$$u(\mathbf{x}) = \int_S \mathcal{G}(\mathbf{x}, \boldsymbol{\xi}; k) p^D(\boldsymbol{\xi}) \, d\Gamma_\xi, \quad \hat{u}(\mathbf{x}) = \int_S \mathcal{G}(\mathbf{x}, \boldsymbol{\xi}; k) \varphi_{,u}(\boldsymbol{\xi}) \, d\Gamma_\xi. \quad (18)$$

It is convenient for the present purposes to split $(\mathcal{G}, \mathcal{H})$ according to:

$$\begin{aligned} \mathcal{G}(\mathbf{x}, \boldsymbol{\xi}; k) &= G(\mathbf{x}, \boldsymbol{\xi}; k) + G_C(\mathbf{x}, \boldsymbol{\xi}; k) \\ \mathcal{H}(\mathbf{x}, \boldsymbol{\xi}; k) &= H(\mathbf{x}, \boldsymbol{\xi}; k) + H_C(\mathbf{x}, \boldsymbol{\xi}; k), \end{aligned} \quad (19)$$

where (G, H) is the well-known singular free-space fundamental solution for the Helmholtz equation, given by

$$G(\mathbf{x}, \boldsymbol{\xi}; k) = \frac{1}{4\pi r} e^{ikr}, \quad H(\mathbf{x}, \boldsymbol{\xi}; k) = [\mathbf{r} \cdot \mathbf{n}(\boldsymbol{\xi})] \frac{ikr - 1}{4\pi r^3} e^{ikr}, \quad (20)$$

with $\mathbf{r} = \boldsymbol{\xi} - \mathbf{x}$ and $r = |\boldsymbol{\xi} - \mathbf{x}| = |\mathbf{r}|$, and the complementary part (G_C, H_C) is not singular at $\boldsymbol{\xi} = \mathbf{x}$. On using decomposition (19), equation (16) becomes

$$\begin{aligned} \frac{1}{2}f(\mathbf{x}) + \int_{\Gamma_\varepsilon} H(\mathbf{x}, \boldsymbol{\xi}; k) v^\varepsilon(\boldsymbol{\xi}) \, d\Gamma_\xi + \int_{\Gamma_\varepsilon} H_C(\mathbf{x}, \boldsymbol{\xi}; k) v^\varepsilon(\boldsymbol{\xi}) \, d\Gamma_\xi \\ = - \int_{\Gamma_\varepsilon} G(\mathbf{x}, \boldsymbol{\xi}; k) p[u](\boldsymbol{\xi}) \, d\Gamma_\xi - \int_{\Gamma_\varepsilon} G_C(\mathbf{x}, \boldsymbol{\xi}; k) p[u](\boldsymbol{\xi}) \, d\Gamma_\xi \quad (\mathbf{x} \in \Gamma_\varepsilon). \end{aligned} \quad (21)$$

3.4 Leading asymptotic contribution to the scattered field

To study the asymptotic behaviour of integral equation (21) as $\varepsilon \rightarrow 0$, it is useful to introduce further scaled geometric quantities:

$$\mathbf{x} = \varepsilon \bar{\mathbf{x}}, \quad \mathbf{r} = \varepsilon \bar{\mathbf{r}}, \quad r = \varepsilon \bar{r} \quad (\mathbf{x}, \boldsymbol{\xi} \in \Gamma_\varepsilon; \bar{\mathbf{x}}, \bar{\boldsymbol{\xi}} \in \mathcal{S}) \quad (22)$$

in addition to definition (4) of $\bar{\boldsymbol{\xi}}$. The leading contributions as $\varepsilon \rightarrow 0$ to the fundamental kernels featured in equation (21) are

$$\begin{aligned} G(\mathbf{x}, \boldsymbol{\xi}; k) &= \varepsilon^{-1} G(\bar{\mathbf{x}}, \bar{\boldsymbol{\xi}}) + O(1) \\ H(\mathbf{x}, \boldsymbol{\xi}; k) &= \varepsilon^{-2} H(\bar{\mathbf{x}}, \bar{\boldsymbol{\xi}}) + O(1) \end{aligned} \quad (\mathbf{x}, \boldsymbol{\xi} \in \Gamma_\varepsilon), \quad (23)$$

for the singular kernels (G, H) defined by (20), where

$$G(\bar{\mathbf{x}}, \bar{\boldsymbol{\xi}}) = \frac{1}{4\pi\bar{r}}, \quad H(\bar{\mathbf{x}}, \bar{\boldsymbol{\xi}}) = -\frac{\bar{\mathbf{r}} \cdot \mathbf{n}(\bar{\boldsymbol{\xi}})}{4\pi\bar{r}^3} \quad (24)$$

are the fundamental kernels for the Laplace equation, and

$$\begin{aligned} G_C(\mathbf{x}, \boldsymbol{\xi}; k) &= G_C(\mathbf{x}_s, \mathbf{x}_s; k) + O(\varepsilon) \\ H_C(\mathbf{x}, \boldsymbol{\xi}; k) &= H_C(\mathbf{x}_s, \mathbf{x}_s; k) + O(\varepsilon) \end{aligned} \quad (\mathbf{x}, \boldsymbol{\xi} \in \Gamma_\varepsilon). \quad (25)$$

for the nonsingular kernels (G_C, H_C) .

On performing the coordinate transformations (4), (22) and invoking estimates (23), (25) together with (5b), one finds that

$$\begin{aligned} \int_{\Gamma_\varepsilon} H_C(\mathbf{x}, \boldsymbol{\xi}; k) v^\varepsilon(\boldsymbol{\xi}) \, d\Gamma_\xi &= O(\varepsilon^2 |v^\varepsilon|) \\ \int_{\Gamma_\varepsilon} G_C(\mathbf{x}, \boldsymbol{\xi}; k) p[u](\boldsymbol{\xi}) \, d\Gamma_\xi &= O(\varepsilon^3). \end{aligned} \quad (26)$$

The limiting form of integral equation (21) as $\varepsilon \rightarrow 0$, retaining only the leading contributions, is hence found to be

$$\bar{\mathcal{L}}_{\mathcal{S}} \left[v^\varepsilon(\mathbf{x}_s + \varepsilon \bar{\boldsymbol{\xi}}) \right] (\bar{\mathbf{x}}) = -\varepsilon \nabla u(\mathbf{x}_s) \cdot \int_{\mathcal{S}} \frac{1}{4\pi\bar{r}} \mathbf{n}(\bar{\boldsymbol{\xi}}) \, d\bar{\Gamma}_{\bar{\boldsymbol{\xi}}} + o(\varepsilon) \quad (27)$$

where $\bar{\mathcal{L}}_{\mathcal{S}}$, defined by

$$\left[\bar{\mathcal{L}}_{\mathcal{S}} f \right] (\bar{\mathbf{x}}) = \frac{1}{2} f(\bar{\mathbf{x}}) + \int_{\mathcal{S}} H(\bar{\mathbf{x}}, \bar{\boldsymbol{\xi}}) f(\bar{\boldsymbol{\xi}}) \, d\bar{\Gamma}_{\bar{\boldsymbol{\xi}}} \quad (\bar{\mathbf{x}} \in \mathcal{S}), \quad (28)$$

is in fact the governing integral operator associated with exterior Neumann problems for the Laplace equation in the normalized domain $\mathbb{R}^3 \setminus \bar{\mathcal{B}}$. Equation (27) indicates that the scattered field is of order $O(\varepsilon)$ on Γ_ε :

$$v^\varepsilon(\boldsymbol{\xi}) = v^\varepsilon(\mathbf{x}_s + \varepsilon \bar{\boldsymbol{\xi}}) = \varepsilon \nabla u(\mathbf{x}_s) \cdot \boldsymbol{\nu}(\bar{\boldsymbol{\xi}}) + o(\varepsilon) \quad (\boldsymbol{\xi} \in \Gamma_\varepsilon, \bar{\boldsymbol{\xi}} \in \mathcal{S}) \quad (29)$$

where the vector function $\boldsymbol{\nu}(\bar{\boldsymbol{\xi}})$ solves the exterior (vector) Laplace problem

$$\left[\bar{\mathcal{L}}_{\mathcal{S}} \boldsymbol{\nu} \right] (\bar{\mathbf{x}}) = -\nabla u(\mathbf{x}_s) \cdot \int_{\mathcal{S}} \frac{1}{4\pi\bar{r}} \mathbf{n}(\bar{\boldsymbol{\xi}}) \, d\bar{\Gamma}_{\bar{\boldsymbol{\xi}}} \quad (30)$$

i.e.

$$\begin{aligned}\Delta \mathbf{V} &= 0 & (\bar{\boldsymbol{\xi}} \in \mathbb{R}^3 \setminus \bar{\mathcal{B}}), \\ \nabla_{\boldsymbol{\xi}} \mathbf{V} \cdot \mathbf{n} &= -\mathbf{n} & (\bar{\boldsymbol{\xi}} \in \mathcal{S}), \\ |\boldsymbol{\xi}|^2 \mathbf{V} &= O(1) & (|\bar{\boldsymbol{\xi}}| \rightarrow +\infty)\end{aligned}\tag{31}$$

It is important to note that $\mathbf{V}(\bar{\boldsymbol{\xi}})$ does not depend on the sampling point \mathbf{x}_s , and hence needs to be computed only once.

3.5 Topological derivative

On substituting (29) into (15) and taking (5) into account, one finally arrives at the following expansion of $J(\varepsilon)$:

$$J(\varepsilon) = J(0) + \varepsilon^3 \mathcal{T}(\mathbf{x}_s) + o(\varepsilon^3)\tag{32}$$

where $\mathcal{T}(\mathbf{x}_s)$, the *topological derivative* of $J(\varepsilon)$, is given in terms of the free and adjoint fields by

$$\mathcal{T}(\mathbf{x}_s) = \text{Re}\left\{\nabla \hat{u} \cdot \mathcal{A}(\mathcal{S}) \cdot \nabla u - |\mathcal{B}| k^2 \hat{u} u\right\}(\mathbf{x}_s)\tag{33}$$

and with the second-order tensor $\mathcal{A}(\mathcal{S})$ defined by

$$\begin{aligned}\mathcal{A}(\mathcal{S}) &= |\mathcal{B}| \mathbf{I} - \int_{\mathcal{S}} \left[\mathbf{V}(\bar{\boldsymbol{\xi}}) \otimes \mathbf{n}(\bar{\boldsymbol{\xi}}) \right] d\bar{\Gamma}_{\bar{\boldsymbol{\xi}}} \\ &= |\mathcal{B}| \mathbf{I} - \int_{\mathcal{S}} \left[\mathbf{n}(\bar{\boldsymbol{\xi}}) \otimes \mathbf{V}(\bar{\boldsymbol{\xi}}) \right] d\bar{\Gamma}_{\bar{\boldsymbol{\xi}}}\end{aligned}\tag{34}$$

where the second equality (i.e. the fact that $\mathcal{A}(\mathcal{S})$ is symmetric) easily stems from the third Green's identity and the definition (31) of \mathbf{V} . The tensor $\mathcal{A}(\mathcal{S})$ is a particular example of the *polarization tensors* that arise in various small-scatterer asymptotic theories, see e.g. [1,2].

For arbitrary surfaces \mathcal{S} which are sufficiently regular for integral equation (30) to be mathematically meaningful (this includes surfaces with edges and corners, e.g. box-shaped scatterers, but precludes infinitely-thin screens), the vector density \mathbf{V} may be found by e.g. solving numerically three sets of BEM equations for exterior Laplace problems, a computationally modest task.

For the simplest case of a rigid spherical obstacle (where \mathcal{B} is the unit sphere, $|\mathcal{B}| = 4\pi/3$, and on which $\mathbf{n}(\bar{\boldsymbol{\xi}}) = -\bar{\boldsymbol{\xi}}$) one easily finds by analytical means that $\mathbf{V}(\bar{\boldsymbol{\xi}}) = \bar{\boldsymbol{\xi}}/2$. Then, (34) is readily found to be given by

$$\mathcal{A}(\mathcal{S}) = 2\pi \mathbf{I}.\tag{35}$$

When \mathcal{S} is an ellipsoid with principal directions $\mathbf{e}'_1, \mathbf{e}'_2, \mathbf{e}'_3$ and semi-axes a_1, a_2, a_3 , the analytical solution of (31) is found by introducing an ellipsoidal

coordinate system and expanding \mathbf{V} in terms of spheroidal harmonics [21]. This procedure yields the polarization tensor featured in (33) as

$$\mathcal{A}(\mathcal{S}) = |\mathcal{B}| \sum_{m=1}^3 \frac{1}{1 - I_m} \mathbf{e}'_m \otimes \mathbf{e}'_m, \quad (36)$$

where I_m are the functions of the two aspect ratios of \mathcal{B} (e.g. a_2/a_1 and a_3/a_1) given by

$$I_m = \frac{a_1 a_2 a_3}{2} \int_0^\infty \frac{dr}{(r + a_m^2) \sqrt{r + a_1^2} \sqrt{r + a_2^2} \sqrt{r + a_3^2}}, \quad m = 1, 2, 3$$

3.6 Qualitative obstacle identification via topological sensitivity

Equation (33) provides, when applied to the featured cost functional \mathcal{J} , an information indicating whether \mathcal{J} increases ($\mathcal{T}(\mathbf{x}_s) > 0$) or decreases ($\mathcal{T}(\mathbf{x}_s) < 0$) in response to the nucleation at \mathbf{x}_s of an infinitely small impenetrable scatterer with prescribed shape. Thus, the topological sensitivity field $\mathcal{T}(\mathbf{x}_s)$ is guaranteed to define a correct obstacle indicator only in the limit as the size of a hidden obstacle approaches zero. It is nonetheless natural to examine whether the spatial distribution of \mathcal{T} can still furnish a useful information for the reconstruction and characterization of *finite* obstacles. In keeping with the ultimate goal of minimizing \mathcal{J} , the key idea in this setting is to approximate the support of a hidden (finite) obstacle via regions where \mathcal{T} attains pronounced negative values. While the reasoning behind such an idea is heuristic, the numerical experiments presented in this article as well as in other related works [18, 5, 13, 12, 23] support the utility of this notion. Since the computation of the *indicator function* \mathcal{T} (over the volume of interest) is significantly faster than an iterative (e.g. minimization-based) inversion, this approach offers a convenient means of extracting preliminary obstacle information from the data u^{obs} . Such information can then be used either in a stand-alone manner for qualitative reconstruction, or as a reliable initial “guess” for more elaborate, iterative reconstruction schemes.

3.7 Practical computation of topological sensitivity

The developments thus far are based on the Green’s function \mathcal{G} defined by (17), and lead to almost explicit formulae for $\mathcal{T}(\mathbf{x}_s)$, their only non-explicit component being the auxiliary density \mathbf{V} , which must be computed numerically except for simple shapes of the trial scatterer \mathcal{B} .

In practice, this explicit character is retained only for geometries Ω such that

the corresponding Green's function is known analytically. Such cases correspond with geometrically simple configurations, e.g. the acoustic half-space. For configurations where the Green's function is not available, the free and adjoint fields and the nonsingular kernels may be sought as solutions of boundary integral equations [4,8]. The free and adjoint fields, defined by (9) and (13), satisfy the well-known integral identities

$$c(\mathbf{x})u(\mathbf{x}) + \int_S H(\mathbf{x}, \boldsymbol{\xi}; k)u(\boldsymbol{\xi}) \, d\Gamma_\xi = \int_S G(\mathbf{x}, \boldsymbol{\xi}; k)p^D(\boldsymbol{\xi}) \, d\Gamma_\xi \quad (37)$$

$$c(\mathbf{x})\hat{u}(\mathbf{x}) + \int_S H(\mathbf{x}, \boldsymbol{\xi}; k)\hat{u}(\boldsymbol{\xi}) \, d\Gamma_\xi = - \int_{S^{\text{obs}}} G(\mathbf{x}, \boldsymbol{\xi}; k)\varphi_{,u}(\boldsymbol{\xi}) \, d\Gamma_\xi \quad (38)$$

which provide integral equations for $\mathbf{x} \in S$ (with $c(\mathbf{x}) = 1/2$, except at points \mathbf{x} where S is only piecewise smooth, such as edges or corners, for which $c(\mathbf{x})$ is also known) and integral representation formulae for $\mathbf{x} \in \Omega$ (with $c(\mathbf{x}) = 1$). Differentiation of the latter under the integral sign provide integral representations of $\nabla u(\mathbf{x})$ and $\nabla u(\mathbf{x}_s)$ in Ω .

Hence, evaluation of the topological sensitivity field as given by (33) entails the solution of integral equations (37) and (38) for u and \hat{u} on S , followed by an evaluation of $u(\mathbf{x}_s)$, $\nabla u(\mathbf{x}_s)$ and $\hat{u}(\mathbf{x}_s)$, $\nabla \hat{u}(\mathbf{x}_s)$ by means of (37) and (38) used as integral representations. The first step may involve large numbers of unknowns if the diameter of Ω spans more than a few wavelengths. Moreover, the present objective being to comprehensively explore a 3-D region of space for hidden scatterers by examining the distribution of $\mathcal{T}(\mathbf{x}_s)$, formula (33) is to be evaluated at a large number of sampling points \mathbf{x}_s . Both steps can then be considerably accelerated using the FM-BEM.

4 The fast multipole method for Helmholtz equation

4.1 BEM discretization

Equations (37) and (38) are in this article solved by means of the simplest BEM discretization, which employs flat triangular boundary elements with straight edges and piecewise-linear C^0 interpolation of u and \hat{u} (other choices, e.g. C^0 quadratic interpolation, would of course have been possible). All numerical results presented hereinafter have been obtained on that basis. The primary unknowns are the values of u or \hat{u} at the mesh nodes, i.e. at all the element vertices. Equations (37) and (38) are collocated at all mesh nodes. All singular element integrals associated with the kernel $H(\mathbf{x}, \boldsymbol{\xi}; k)$ are zero because of the piecewise-flat geometry representation, which of course simplifies the implementation. Denoting by N the total number of nodes (and hence of

unknowns), this procedure gives rise to the linear systems of equations

$$[A]\{u\} = \{b\} \quad (39)$$

$$[A]\{\hat{u}\} = \{\hat{b}\} \quad (40)$$

where the N -vectors $\{u\}$ and $\{\hat{u}\}$ collect all nodal values of u and \hat{u} . The discussion to follow will focus on solving system (39), the adjoint system (40) being of course solved in exactly the same way with $\{b\}$ replaced by $\{\hat{b}\}$.

As the problem size N grows, direct solvers become impractical or infeasible with respect to both computing time and storage, mainly due to the fully-populated nature of the BEM matrix $[A]$, and iterative solvers are used instead. Since $[A]$ is a non-symmetric, invertible matrix (except when k is an eigenvalue for Ω and homogeneous Neumann BCs), the iterative solution technique most often used for systems such as (39) is the generalized minimal residual (GMRES) algorithm (see e.g. [15]), which is applicable to general invertible square matrices. Such algorithms are based on matrix-vector evaluations, and therefore do not require actual assembly and storage of $[A]$. The GMRES algorithm requires repeated evaluations of the *residual*

$$\{b\} - [A]\{u\} \quad (41)$$

where $\{u\}$ is a given solution candidate, which is updated at each GMRES iteration. Hence, one needs to compute (discretized versions of) the double-layer and single-layer potentials featured in the left-hand and right-hand sides, respectively, of (37) and (38) for known densities.

Traditional BE methods lead to a $O(N^2)$ computational cost for each residual evaluation, because element integrals computed for a collocation point cannot be reused for another collocation point. By adopting the Fast Multipole boundary element method (FM-BEM), each residual can be computed within a $O(N \log N)$ time. The implementation used here, concisely described in the remainder of this section, follows Darve [9] and Sylvand [31,32].

4.2 Expansion of the fundamental solution

The starting point for the FM-BEM is the following representation of the full-space fundamental solution (20):

$$G(\mathbf{x}, \boldsymbol{\xi}; k) = \lim_{p \rightarrow \infty} \frac{ik}{4\pi} \int_{\hat{S}} e^{-ik(\hat{\mathbf{s}} \cdot \tilde{\mathbf{x}})} T_p(\hat{\mathbf{s}}, \mathbf{r}_0) e^{ik(\hat{\mathbf{s}} \cdot \tilde{\boldsymbol{\xi}})} d\Gamma_{\hat{\mathbf{s}}} \quad (42)$$

where $\hat{S} = \{\hat{\mathbf{s}}, |\hat{\mathbf{s}}| = 1\}$ is the unit sphere, the position vector $\mathbf{r} = \boldsymbol{\xi} - \mathbf{x}$ has been decomposed as

$$\mathbf{r} = (\boldsymbol{\xi}_0 - \mathbf{x}_0) + (\boldsymbol{\xi} - \boldsymbol{\xi}_0) - (\mathbf{x} - \mathbf{x}_0) = \mathbf{r}_0 + \tilde{\boldsymbol{\xi}} - \tilde{\mathbf{x}} \quad (43)$$

in terms of two poles \mathbf{x}_0 and $\boldsymbol{\xi}_0$, and with the transfer function $T_p(\hat{\mathbf{s}}, \mathbf{r}_0)$ defined by

$$T_p(\hat{\mathbf{s}}, \mathbf{r}_0) = \frac{(-i)^n}{4\pi} \sum_{n=0}^p (-1)^n (2n+1) h_n^{(1)}(kr_0) P_n(\hat{\mathbf{s}} \cdot \hat{\mathbf{r}}_0) \quad (44)$$

In (44), the $h_n^{(1)}$ and P_n are respectively the spherical Hankel functions of first kind and the Legendre polynomials, and $r_0 \equiv |\boldsymbol{\xi}_0 - \mathbf{x}_0|$. Moreover, for any vector $\mathbf{z} \in \mathbb{R}^3$, a hat symbol indicates the corresponding unit vector, i.e. $\hat{\mathbf{z}} = \mathbf{z}/|\mathbf{z}|$. Representation (42) holds under the condition

$$|r_0| > |\tilde{\boldsymbol{\xi}} - \tilde{\mathbf{x}}| \quad (45)$$

The poles \mathbf{x}_0 and $\boldsymbol{\xi}_0$ are actually meant to be chosen close to $\boldsymbol{\xi}$ and \mathbf{x} , respectively, so as to satisfy the stronger condition

$$|\boldsymbol{\xi} - \mathbf{x}_0| > |\mathbf{x} - \mathbf{x}_0| \quad \text{and} \quad |\mathbf{x} - \boldsymbol{\xi}_0| > |\boldsymbol{\xi} - \boldsymbol{\xi}_0| \quad (46)$$

The representation of $H(\mathbf{x}, \boldsymbol{\xi}; k)$ obtained by differentiating (42) is then

$$H(\mathbf{x}, \boldsymbol{\xi}; k) = \lim_{p \rightarrow \infty} \frac{-k^2}{4\pi} \int_{\hat{S}} e^{-ik(\hat{\mathbf{s}} \cdot \tilde{\mathbf{x}})} T_p(\hat{\mathbf{s}}, \mathbf{r}_0) [\hat{\mathbf{s}} \cdot \mathbf{n}(\boldsymbol{\xi})] e^{ik(\hat{\mathbf{s}} \cdot \tilde{\boldsymbol{\xi}})} d\Gamma_{\hat{\mathbf{s}}} \quad (47)$$

In practice, representations (42) and (47) are approximated by (i) using the transfer function $T_p(\hat{\mathbf{s}}, \mathbf{r}_0)$ with a finite index p and (ii) replacing the integral over the unit sphere with a quadrature rule with Q points $\hat{\mathbf{s}}_q$ and weights w_q . So, one replaces (42) and (47) with the approximations

$$G(\mathbf{x}, \boldsymbol{\xi}; k) \approx \frac{ik}{4\pi} \sum_{q=1}^Q w_q e^{-ik(\hat{\mathbf{s}}_q \cdot \tilde{\mathbf{x}})} T_p(\hat{\mathbf{s}}_q, \mathbf{r}_0) e^{ik(\hat{\mathbf{s}}_q \cdot \tilde{\boldsymbol{\xi}})} \quad (48)$$

$$H(\mathbf{x}, \boldsymbol{\xi}; k) \approx \frac{ik}{4\pi} \sum_{q=1}^Q w_q e^{-ik(\hat{\mathbf{s}}_q \cdot \tilde{\mathbf{x}})} T_p(\hat{\mathbf{s}}_q, \mathbf{r}_0) [\hat{\mathbf{s}}_q \cdot \mathbf{n}(\boldsymbol{\xi})] e^{ik(\hat{\mathbf{s}}_q \cdot \tilde{\boldsymbol{\xi}})} \quad (49)$$

The choice of points $\hat{\mathbf{s}}_q$ and weights w_q , and their number Q , depends on the truncation order p used in the transfer function T_p . On parameterizing unit vectors $\hat{\mathbf{s}} \in \hat{S}$ using spherical angular coordinates (θ, ϕ) , a commonly used choice [31,9] consists in using the $Q(p) = (p+1)(2p+1)$ quadrature points of the form $\hat{\mathbf{s}}_q = (\theta_a, \phi_b)$, where $\cos \theta_a$ ($1 \leq a \leq p+1$) are the abscissae for the $(p+1)$ -point Gauss-Legendre 1-D quadrature rule over $[-1, 1]$, $\phi_b = 2\pi b/(2p+1)$ ($1 \leq b \leq 2p+1$) are uniformly-spaced abscissae on $[0, 2\pi]$. The associated weights are $w_q = 2\pi w_a^\theta/(2p+1)$, where w_a^θ are the Gauss-Legendre weights for the $(p+1)$ -point rule.

Now, let S_x and S_ξ denote two disjoint portions of S , and let the poles \mathbf{x}_0 and ξ_0 be chosen close to S_x and S_ξ , respectively, in such a way that (46) holds for any $\mathbf{x} \in S_x$ and $\xi \in S_\xi$. Consider the computation of

$$I(\mathbf{x}) = \int_{S_\xi} G(\mathbf{x}, \xi; k) v(\xi) d\Gamma_\xi \quad (\mathbf{x} \in S_x)$$

for a given density $v(\xi)$, which is a typical contribution to the evaluation of the residual (41). On substituting (48), one obtains

$$I(\mathbf{x}) \approx \frac{ik}{4\pi} \sum_{q=1}^{Q(p)} w_q e^{-ik(\hat{\mathbf{s}}_q \cdot \tilde{\mathbf{x}})} T_p(\hat{\mathbf{s}}_q, \mathbf{r}_0) \int_{S_\xi} e^{ik(\hat{\mathbf{s}}_q \cdot \tilde{\xi})} v(\xi) d\Gamma_\xi \quad (\mathbf{x} \in S_x) \quad (50)$$

So, the *same* integral over S_ξ can be re-used for *all* collocation points $\mathbf{x} \in S_x$. Moreover, for a chosen portion S_ξ , this is true for any portion S_x and associated pole \mathbf{x}_0 such that condition (46) holds. Computations of the form (50) can be decomposed into three stages: (i) compute for each quadrature point of \hat{S} the multipole moment $\mathcal{R}(\hat{\mathbf{s}}_q; \xi_0)$:

$$\mathcal{R}(\hat{\mathbf{s}}_q; \xi_0) = \int_{S_\xi} e^{ik(\hat{\mathbf{s}}_q \cdot \tilde{\xi})} v(\xi) d\Gamma_\xi \quad (51)$$

(ii) multiply $\mathcal{R}(\hat{\mathbf{s}}_q; \xi_0)$ by the transfer function $T_p(\hat{\mathbf{s}}_q, \mathbf{r}_0)$, to obtain local expansion coefficients $\mathcal{L}(\hat{\mathbf{s}}_q; \mathbf{x}_0)$ at \mathbf{x}_0 :

$$\mathcal{L}(\hat{\mathbf{s}}_q; \mathbf{x}_0) = T_p(\hat{\mathbf{s}}_q, \xi_0 - \mathbf{x}_0) \mathcal{R}(\hat{\mathbf{s}}_q; \xi_0) \quad (52)$$

(iii) for all $\mathbf{x} \in S_x$, transfer $\mathcal{L}(\hat{\mathbf{s}}_q; \mathbf{x}_0)$ locally from \mathbf{x}_0 to \mathbf{x} and perform the numerical quadrature over \hat{S} , to obtain (an approximation of) $I(\mathbf{x})$:

$$I(\mathbf{x}) \approx \frac{ik}{4\pi} \sum_{q=1}^{Q(p)} w_q e^{-ik(\hat{\mathbf{s}}_q \cdot \tilde{\mathbf{x}})} \mathcal{L}(\hat{\mathbf{s}}_q; \mathbf{x}_0) \quad (53)$$

The one-level fast multipole method consists in partitioning the spatial region containing S into cubic cells of identical sizes whose vertices lie on a regular cubic grid. Each pair (S_x, S_ξ) is such that $S_x = S \cap C_x$ and $S_\xi = S \cap C_\xi$, where (C_x, C_ξ) is any pair of *disjoint* cubic cells. The poles \mathbf{x}_0 and ξ_0 are the respective cell centroids. The one-level FM-BEM has a complexity of $O(N^{3/2})$ per iteration for equations of type (39), which is of course better than the $O(N^2)$ complexity of traditional BEM, but not optimal. Further acceleration is provided by using the multi-level fast multipole method, where the size of clusters S_x, S_ξ depends of their distance.

4.3 Multilevel FM-BEM algorithm

To exploit optimally the acceleration afforded by (50), a hierarchical oct-tree structure of elements is introduced. For that purpose, a cube containing the boundary S , called ‘level-0 cell’, is divided into eight cubes (level-1 cells), each of which is divided in the same fashion, and so on. A level- ℓ cell is divided into level- $(\ell+1)$ cells unless it contains less than a preset (relatively small) number E of boundary elements (such cells are termed *leaves*). A noteworthy feature of the FM-BEM applied to Helmholtz-type equations is that to achieve the same accuracy in approximations (48), (49) at all levels, the truncation parameter p is level-dependent. A often-used formula [9] for the adjustment of p is of the form

$$p(\ell) = \sqrt{3}kd(\ell) + C\text{Log}_{10}(\sqrt{3}kd(\ell) + \pi) \quad (54)$$

where $d(\ell)$ is the size of a cubic ℓ -level cell (so, $d(\ell+1) = 2d(\ell)$ and $p(\ell+1)$ is roughly $2p(\ell)$) and C is a constant. As a result, the set of quadrature points $\hat{\mathbf{s}}_q$ on \hat{S} is also level-dependent: each level necessitates a distinct set $\hat{\mathbf{s}}_q^\ell$ of quadrature points and associated weights. For the present implementation, values of C such that $2 \leq C \leq 8$ were found to provide a acceptable compromise between accuracy and cost, and $C = 4$ was actually used.

The FM-BEM algorithm then consists of:

- An *upward pass* where multipole moments (51) are first computed for the lowest-level cells and then recursively aggregated by moving upward in the tree until level 2 (for which there are $4 \times 4 \times 4$ cells) is reached. Letting $\mathcal{C}(\boldsymbol{\xi}_0)$ denote the set of children of a given ℓ -level cell $C(\boldsymbol{\xi}_0)$, i.e. of $(\ell+1)$ -level cells $C(\boldsymbol{\xi}'_0)$ contained in $C(\boldsymbol{\xi}_0)$, this operation relies on the identity

$$\mathcal{R}(\hat{\mathbf{s}}_q^{\ell+1}; \boldsymbol{\xi}_0) = \sum_{C(\mathbf{y}_0) \in \mathcal{C}(\mathbf{x}_0)} e^{ik(\hat{\mathbf{s}}_q^{\ell+1} \cdot (\boldsymbol{\xi}'_0 - \boldsymbol{\xi}_0))} \mathcal{R}(\hat{\mathbf{s}}_q^{\ell+1}; \boldsymbol{\xi}'_0) \quad (55)$$

for shifting the origin from the center $\boldsymbol{\xi}'_0$ of a level- $(\ell+1)$ cell to the center $\boldsymbol{\xi}_0$ of a level- ℓ cell in the the contribution of a given cell.

Then, it is necessary to interpolate, i.e. compute $\mathcal{R}(\hat{\mathbf{s}}_q^\ell; \boldsymbol{\xi}_0)$ at the quadrature points $\hat{\mathbf{s}}_q^\ell$ from the previously determined values $\mathcal{R}(\hat{\mathbf{s}}_q^{\ell+1}; \boldsymbol{\xi}_0)$. The procedure used follows [31,9] and is not detailed here.

- A *downward pass* where the coefficients of local expansions are first computed at level $\ell = 2$ and then evaluated at lower-level cells by tracing the tree structure downwards. This operation relies on the identity

$$\mathcal{L}(\hat{\mathbf{s}}_q^\ell; \mathbf{x}'_0) = e^{-ik(\hat{\mathbf{s}}_q^{\ell+1} \cdot (\mathbf{x}_0 - \mathbf{x}'_0))} \mathcal{L}(\hat{\mathbf{s}}_q^\ell; \mathbf{x}_0) \quad (56)$$

for shifting the origin from the center \mathbf{x}_0 of a level- (ℓ) cell to the center \mathbf{x}'_0

of a level- $(\ell+1)$ cell. This operation is not performed at the root level, i.e. when $\ell+1=2$.

Then, the contributions of all level- $(\ell+1)$ cells belonging to the *interaction list* $\mathcal{I}(\mathbf{x}_0)$ of the level- $\ell+1$ cell $C(\mathbf{x}_0)$ (i.e. all such cells which are non-adjacent to $C(\mathbf{x}_0)$ but whose father is adjacent to the father of $C(\mathbf{x}_0)$) are aggregated:

$$\sum_{C(\mathbf{y}_0) \in \mathcal{I}(\mathbf{x}_0)} T_p(\hat{\mathbf{s}}_q^\ell, \mathbf{y}_0 - \mathbf{x}_0) \mathcal{R}(\hat{\mathbf{s}}_q^\ell; \mathbf{y}_0)$$

and the result is added to $\mathcal{L}(\hat{\mathbf{s}}_q^\ell; \mathbf{x}_0')$ given by (56)

Then, the values $\mathcal{L}(\hat{\mathbf{s}}_q^\ell; \mathbf{x}_0')$ are converted to values at the quadrature points $\hat{\mathbf{s}}_q^{\ell+1}$ by a ‘reverse interpolation’, or ‘anterpolation’, procedure.

- When the lowest level is reached, all quadratures of the form (53) are finally performed, where \mathbf{x}_0 is the centers of a leaf cell, thus evaluating all far-field contributions to the residual at all collocation points.

Moreover, for all leaf cells $C(\mathbf{x}_0)$ and all collocation points $\mathbf{x} \in C(\mathbf{x}_0)$, the near-field contributions are computed by evaluating using traditional integration methods the element contributions for all elements located in $C(\mathbf{x}_0)$ and all cells of same level adjacent to $C(\mathbf{x}_0)$.

The computation of integral representation formulae for $u(\mathbf{x}_s)$, $\nabla u(\mathbf{x}_s)$ and $\hat{u}(\mathbf{x}_s)$, $\nabla \hat{u}(\mathbf{x}_s)$ at all chosen sampling points follows the same approach, with collocation points $\mathbf{x} \in S$ replaced with sampling points $\mathbf{x}_s \in \Omega$. For sampling points lying in leaf cells not adjacent to any same-level cell intersecting S , the integral representations result from far-field interactions (i.e. fast-multipole contributions) only. Besides, all multipole moments used in this step are those corresponding to the solution of (39) or (40), i.e. those evaluated at the last iteration of the GMRES solution algorithm

4.4 Numerical verification of theoretical complexity

To check that the theoretical complexity of the FM-BEM is achieved and illustrate the computational advantage brought by the FM-BEM over the conventional BEM, the simple situation of a spherical acoustic domain subjected on its surface to a uniform normal velocity, is considered. BEM solutions for this problem have been computed for a sequence of meshes with decreasing element size. For each mesh, the prescribed frequency is selected so that the number of nodes per wavelength is (approximately) the same for all meshes. Figure 1 shows the CPU time used for one evaluation of residual (41), using either classical BEM integration techniques (with theoretical complexity $O(N^2)$) or the FM-BEM (with theoretical complexity $N \log N$), as a function of the number N of nodal unknowns. Functions of the form $C_1 N^2$ and $C_2 N \log N$ closest to the actual recorded values of $\text{CPU}(N)$ are also plotted on the same graph. The theoretical complexity for both the classical BEM and the FM-BEM are very

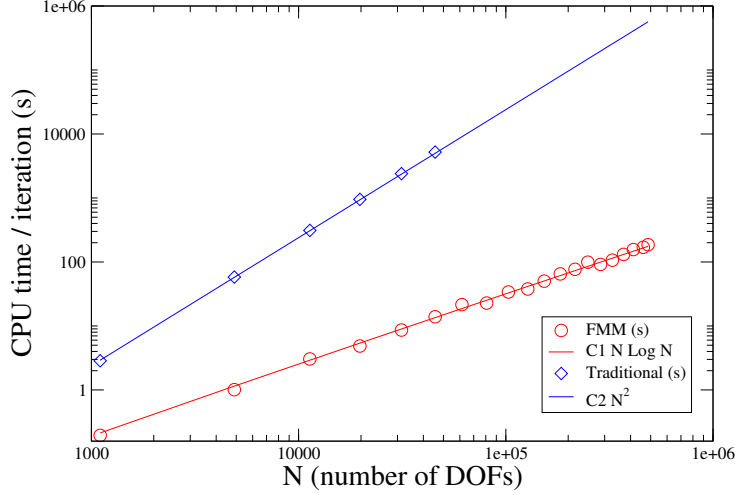


Fig. 1. CPU time for one evaluation of residual (41) as a function of the number N of nodal unknowns, with matrix-vector products computed using classical BEM techniques (lozenges, theoretical complexity $O(N^2)$ or the FM-BEM (circles, theoretical complexity $O(N\log N)$).

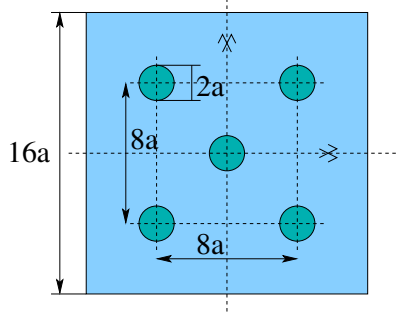


Fig. 2. Pattern of excitation surfaces S_q on each face of external boundary S .

well verified by the actual CPU times. The FM-BEM, as expected, performs much better for large problems.

5 Preliminary identification via topological sensitivity: numerical examples

To illustrate the approach described in Sections 3 and 4, the following configuration has been considered: the bounded acoustic domain is the cube defined by $\Omega(L) = \{ |\xi_i| \leq L \ (i = 1, 2, 3) \}$, with $L = 8a$ or $L = 16a$ in terms of a reference length a . A simulated testing configuration is based on 30 experiments, each of which consists in applying a uniform excitation over a small region S_q ($1 \leq q \leq 30$) of S and recording the acoustic pressure over the whole boundary S (i.e. at all BE mesh nodes), so that $S^{\text{obs}} = S$. Unless indicated otherwise, the acoustic excitation is such that the wavelength is $\lambda = 3a$. Each of the six

faces of the cubical domain Ω supports five excitation surfaces S_q , each a disk of radius a , arranged as depicted on Fig. 2. The cost function for the inverse problem is defined by

$$\mathcal{J}(\Omega^*) = \frac{1}{2} \sum_{q=1}^{30} \int_S |u^* - u_q^{\text{obs}}|^2 \, d\Gamma_\xi \quad (57)$$

where u_q^{obs} is the data obtained for the q -th applied excitation, with $u_q^{\text{obs}} = u^{\text{true}}$ in the absence of data noise. The centroid \mathbf{x}^{true} of true scatterer B^{true} to be identified is located at $\mathbf{x}^{\text{true}} = (2a, 3a, 2a)$.

To facilitate the graphical interpretation, a thresholded variant $\hat{\mathcal{T}}(\mathbf{x}_s)$ of $\mathcal{T}(\mathbf{x}_s)$ is introduced according to

$$\hat{\mathcal{T}}(\mathbf{x}_s) = \begin{cases} \mathcal{T}(\mathbf{x}_s), & \mathcal{T} \leq C \mathcal{T}_{\min} \\ 0, & \mathcal{T} > C \mathcal{T}_{\min} \end{cases} \quad \text{with } C = 0.25. \quad (58)$$

The BE meshes used for computing the free field u , the adjoint field \hat{u} and the simulated data u_q^{true} are made of three-noded flat triangular elements, arranged in a regular mesh with approximately 15 nodes per wavelength. For the purposes of computing the simulated error-free data u^{true} for each synthetic experiment, BE meshes of Γ^{true} have been set up as well. Table 1 indicates the numbers of nodes and elements supported by the BE meshes.

Table 1

Number of element and DOFs supported by the BE meshes.

Cube size	Cube		Obstacle		Total	
	Elements	nodes	Elements	nodes	Elements	nodes
$2L = 16a$	76800	38402	336	170	77136	38572
$2L = 32a$	307200	153602	336	170	307536	153772

Single spherical scatterer. First, the identification of one spherical scatterer is considered, for three cases with increasing scatterer radii $0.2a$, $0.4a$ and $0.8a$. The field \mathcal{T} has been computed for each case on the basis of error-free synthetic data, over a sampling grid \mathcal{S} made of $100 \times 100 \times 100$ sampling points located on the vertices of a regular cubic grid, centered at the origin and with grid spacing $\Delta x_s = 16a/101$, uniformly filling the whole acoustic domain bounded by S . Figures 3, 4 and 5 show, for each scatterer configuration considered in turn, the distribution of the thresholded topological sensitivity $\hat{\mathcal{T}}$ defined by (58) in the three coordinate planes containing the true scatterer centroid \mathbf{x}^{true} . In all cases, $\hat{\mathcal{T}}$ (and hence \mathcal{T}) is seen to attain its lowest values in zones corresponding to, or close to, the actual true scatterer location. It should however be mentioned that \mathcal{T} has been observed to also achieve low values

in regions close to edges or corners of S , where there is no scatterer. To emphasize and illustrate this remark, Fig. 6 shows the iso-surfaces of the field \mathcal{T} corresponding to $\mathcal{T} = 0.55\mathcal{T}_{\min}$, computed on the entire search grid \mathcal{S} (Fig. 6a) and on truncated search grids such that $\{|\xi_i| \leq 6.5a \ (i = 1, 2, 3)\}$ (Fig. 6b) and $\{|\xi_i| \leq 5a \ (i = 1, 2, 3)\}$ (Fig. 6c), i.e. in which sampling points situated at distances less than $1.5a$ and $3a$, respectively, from S have been taken out of \mathcal{S} . Figure 6c shows that, at least in the central region $\{|\xi_i| \leq 5a \ (i = 1, 2, 3)\}$ of the acoustic domain, low negative values of \mathcal{T} occur only in a small region which is consistent with the actual location of the scatterer. If a lower excitation frequency is used, the distribution of \mathcal{T} around the true obstacle location is more “smeared”, as in Figure 7 where $\lambda = 12a$ was used (the other data being as in the previously discussed examples).

Dual spherical scatterer. The identification of a set of two spherical scatterers, of respective locations $\mathbf{x}_1^{\text{true}} = (a, 2a, 2a)$ and $\mathbf{x}_2^{\text{true}} = (-3a, -2a, 2a)$ and of equal radii $0.8a$, embedded in the same domain $\Omega(8a)$ and under the same (virtual) experimental conditions as before, is now considered. Figure 8, which shows the distribution of thresholded topological sensitivity $\hat{\mathcal{T}}$ in the coordinate planes $\xi_3 = 2a$ containing the true scatterer centroids $\mathbf{x}_{1,2}^{\text{true}}$, indicates that both scatterers are well resolved and reasonably located. Again, \mathcal{T} is observed to also achieve relatively low values in regions close to edges or corners of S , where there is no scatterer. However, as evidenced by a comparison of Figure 8a and 8b, which differ only by the chosen threshold ($C = 0.25$ and $C = 0.5$, respectively) in the definition (58) of $\hat{\mathcal{T}}$, the lowest values of \mathcal{T} are attained near the correct obstacle location, and not near edges or corners of S . These observations are emphasized by Fig. 9 showing the iso-surface of the field \mathcal{T} corresponding to $\mathcal{T} = 0.75\mathcal{T}_{\min}$, computed on a truncated search grids such that $\{|\xi_i| \leq 7a \ (i = 1, 2, 3)\}$. Again, the two obstacles are seen to be well resolved, with no other significantly low value of \mathcal{T} found in the sampling region of interest $\{|\xi_i| \leq 7a \ (i = 1, 2, 3)\}$.

Single spherical scatterer, effect of data noise. Then, the effect of data errors is considered for the spherical scatterer of radius $0.8a$, by using synthetic data u^{obs} in the form $u^{\text{obs}} = u^{\text{true}}(1 + \eta)$, where η are random numbers with zero mean and uniform distribution over the interval $[-0.1, 0.1]$. Figure 10 shows the distribution of $\hat{\mathcal{T}}$ in the three coordinate planes containing \mathbf{x}^{true} . A comparison between Figs. 5 and 10 reveals that the distribution of $\hat{\mathcal{T}}$ is only marginally affected by the data noise.

Box-shaped scatterer. The examples shown thus far illustrate the capability of \mathcal{T} , here defined on the basis of an asymptotic analysis involving vanishing *spherical* obstacles, to identify the location of obstacles of finite extent

having the same assumed shape. Now, the identification of a non-spherical, box-shaped obstacle whose sides are aligned along the coordinate axes and of finite size $0.8a \times 0.8a \times 1.6a$ and whose centroid is still $\mathbf{x}^{\text{true}} = (2a, 3a, 2a)$, is considered. Figure 11 shows the distribution of $\hat{\mathcal{T}}$ in the three coordinate planes containing \mathbf{x}^{true} . The true obstacle is again satisfactorily located.

Single spherical scatterer, larger background domain. A last example illustrates the case of a larger acoustic domain $\Omega(16a)$, instead of $\Omega(8a)$ considered up to this point, with the same wavelength $\lambda = 3a$ as before. The ‘true’ scatterer (again a sphere of radius $0.8a$) is still located at $\mathbf{x}^{\text{true}} = (2a, 3a, 2a)$, and hence is located at a larger distance (expressed in wavelengths) from the measurement surfaces. The sampling grid \mathcal{S} is now made of $150 \times 150 \times 150$ regularly-spaced sampling points, with a grid spacing now of $\Delta x_s = 32a/151$. Figure 12 shows the distribution of $\hat{\mathcal{T}}$ in the three coordinate planes containing \mathbf{x}^{true} . Presumably as a result of greater remoteness (and hence lower identifiability) of the scatterer, these distributions show, in addition to the correct one, secondary spatial zones where the presence of a small scatterer is also consistent with the data. Still, the lowest values of $\hat{\mathcal{T}}$ furnish a reasonable indication of the true obstacle location, as seen on the 3-D plots of iso-surfaces $\mathcal{T} = 0.55\mathcal{T}_{\min}$ and $\mathcal{T} = 0.7\mathcal{T}_{\min}$ of Fig. 13 for the truncated grid defined by $\{|\xi_i| \leq 14a \ (i = 1, 2, 3)\}$. Again, one notes that values of \mathcal{T} close to its minimum \mathcal{T}_{\min} occur only in the vicinity of the correct obstacle location (excluding, as before, regions close to the external surface S). From the 3-D plots of Figure 14, similar remarks apply for the detection of a larger obstacle, of radius $1.2a$, in otherwise identical conditions.

Computational efficiency. Typical CPU times and GMRES iteration counts are provided in Table 2. It is interesting to observe that the overall CPU times for $\Omega(16a)$, which involves roughly 4 times as many nodal unknowns as $\Omega(8a)$, are about 5 times higher than those for $\Omega(8a)$ (while the expected ratio for traditional BEM would be $4^3 = 64$), and that the GMRES iteration counts are only fractionally higher for $\Omega(16a)$. All computations have been performed on a Linux PC computer with one 3 GHz processor.

Table 2

CPU times and (in parentheses) GMRES iteration count for computing the true, free and adjoint solutions on the boundary, and CPU times for computing the topological sensitivity over the whole sampling grid \mathcal{S} .

	u^{true} on $S \cup \Gamma^{\text{true}}$	u on S	\hat{u} on S	\mathcal{T} on \mathcal{S}
$2L = 16a$	1444s (435)	969s (282)	1163s (342)	852s
$2L = 32a$	6461s (439)	5615s (388)	6818s (476)	1860s

6 Conclusion

In this article, a computationally fast qualitative technique for probing acoustic media for hidden hard obstacles on the basis of overdetermined boundary data, based on the computation via the FM-BEM of the distribution of topological sensitivity of the cost functional associated with the inverse problem, has been presented. Its usefulness has been demonstrated through results of numerical experiments on the qualitative identification of hard obstacles in a bounded acoustic domain, for configurations featuring $O(10^5)$ nodal unknowns and $O(10^6)$ sampling points, resulting in overall computing times of a few hours on a 3 GHz PC computer. There is ample scope for increasing these computational sizes and further enhancing computational efficiency. Besides, the proposed approach can be developed for many other physical models, e.g. elastodynamics or electromagnetic waves.

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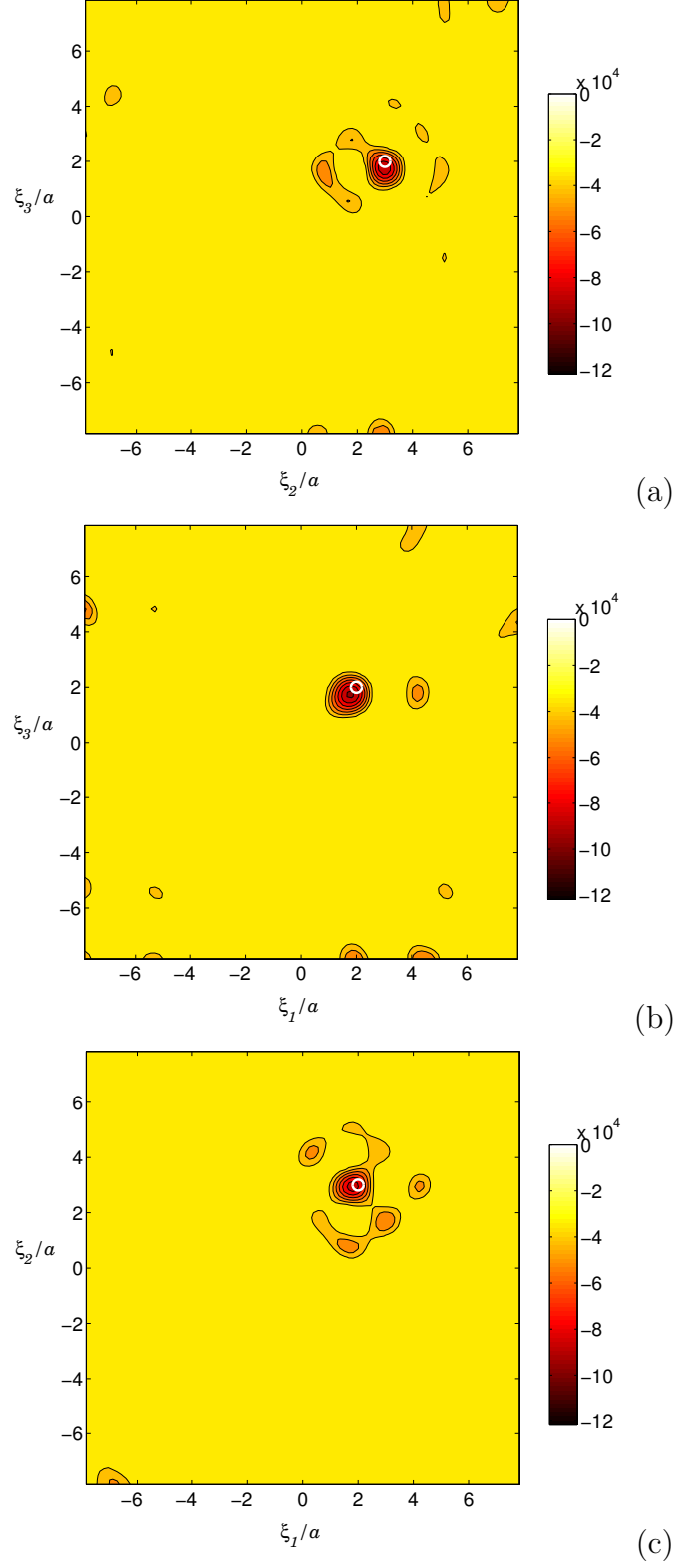


Fig. 3. Identification of spherical hard scatterer of radius $0.2a$: distribution of thresholded topological sensitivity $\hat{\mathcal{T}}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

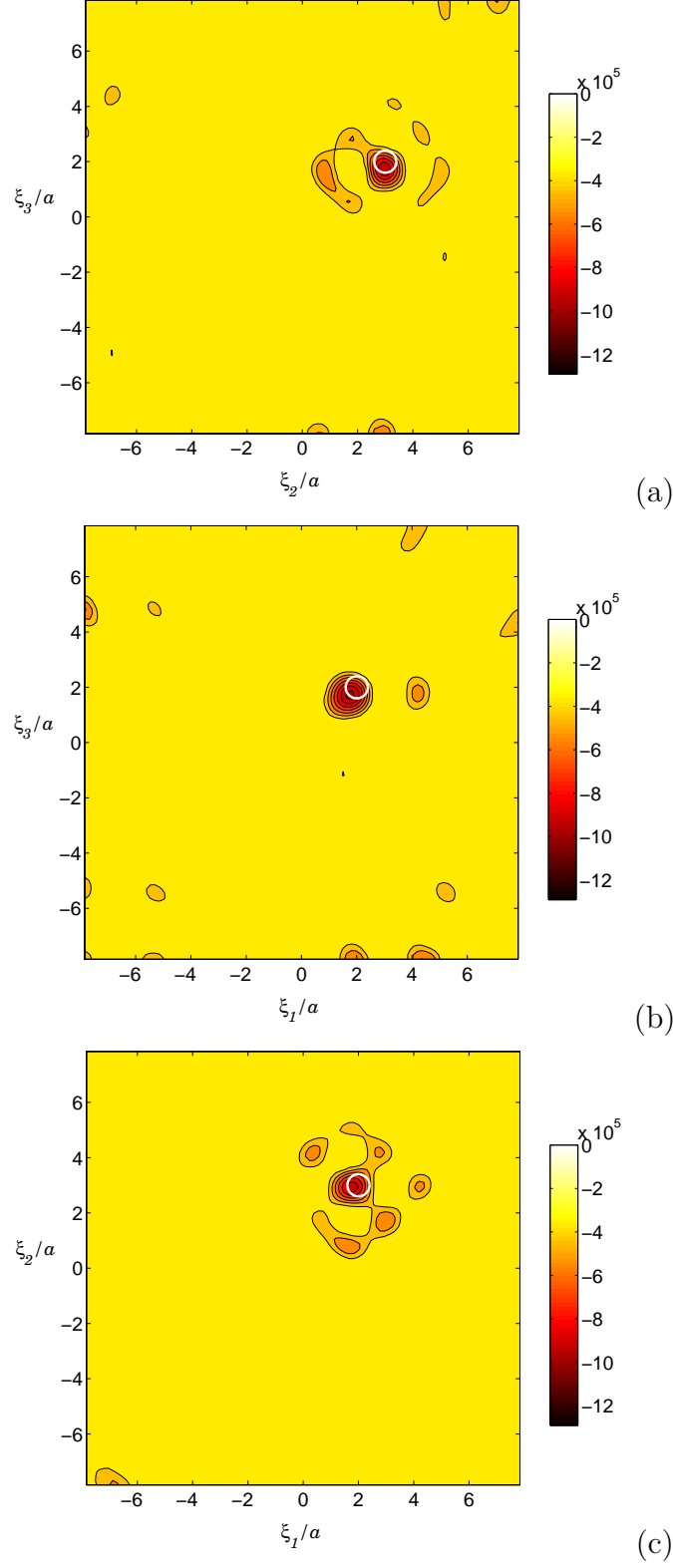


Fig. 4. Identification of spherical hard scatterer of radius $0.4a$: distribution of thresholded topological sensitivity $\hat{\mathcal{T}}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

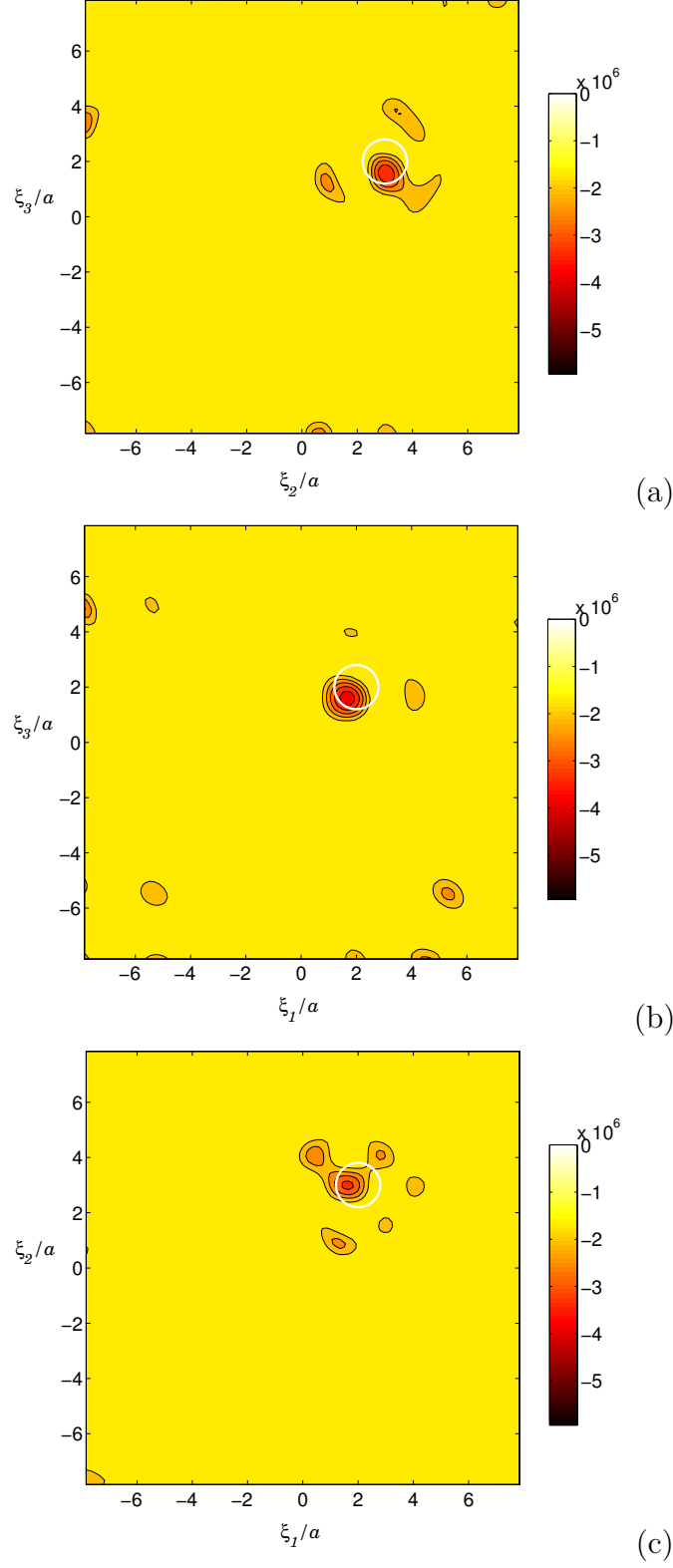


Fig. 5. Identification of spherical hard scatterer of radius $0.8a$: distribution of thresholded topological sensitivity $\hat{\mathcal{T}}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

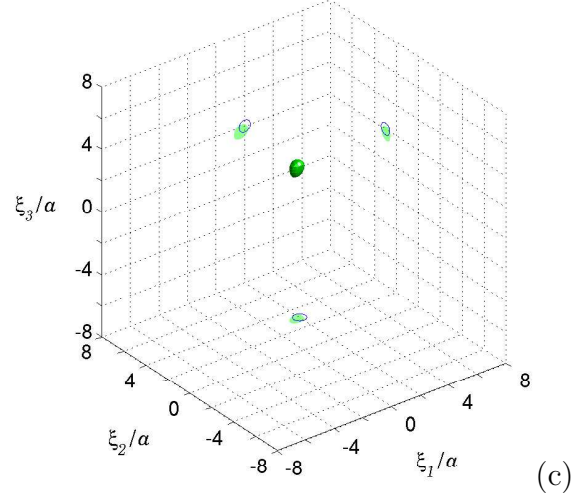
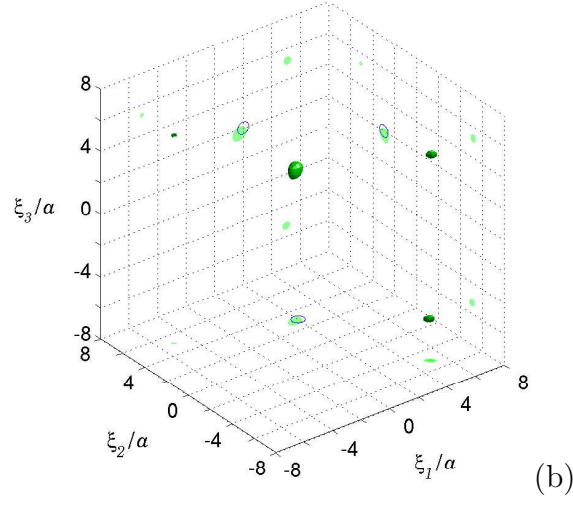
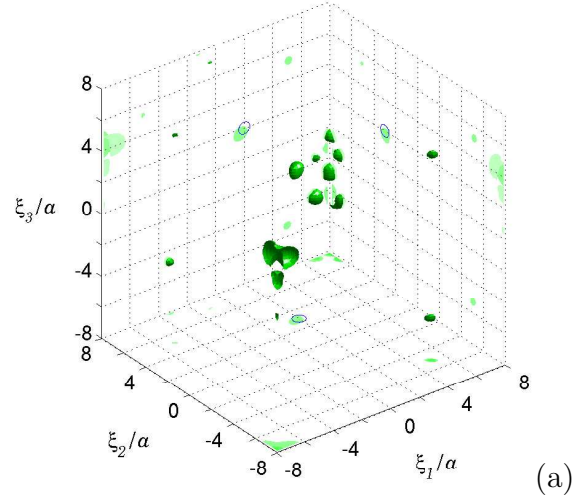


Fig. 6. Identification of spherical hard scatterer of radius $0.4a$: iso-surfaces of $\mathcal{T}(\mathbf{x}_s)$ for $\mathcal{T} = 0.55\mathcal{T}_{\min}$, computed on the entire search grid \mathcal{S} (a) and on truncated search grids such that $\{|\xi_i| \leq 6.5a \ (i=1,2,3)\}$ (b) and $\{|\xi_i| \leq 5a \ (i=1,2,3)\}$ (c). Values of $\mathcal{T}(\mathbf{x}_s)$ lower than the iso-value are inside the iso-surface.

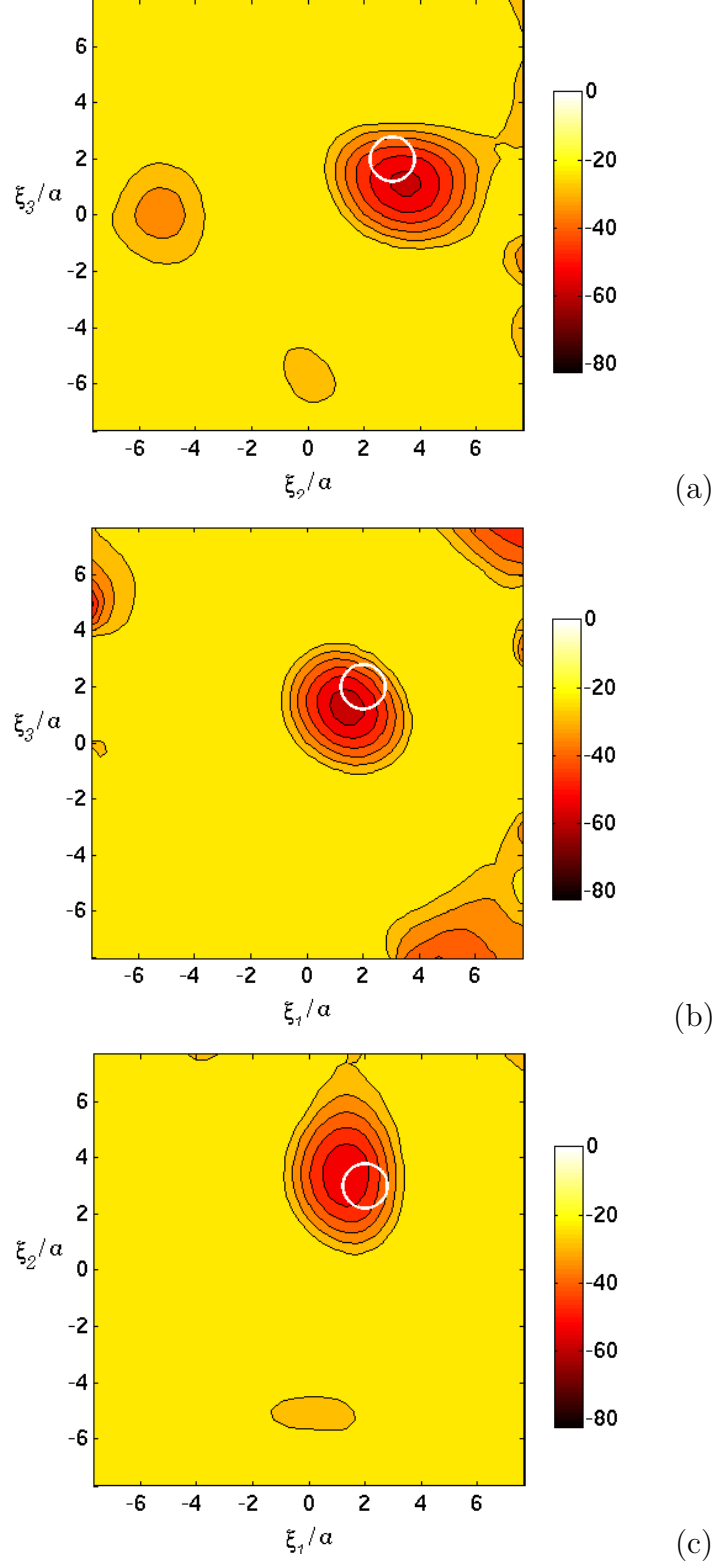


Fig. 7. Identification of spherical hard scatterer of radius $0.8a$, low-frequency excitation ($\lambda = 12a$): distribution of thresholded topological sensitivity $\hat{T}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

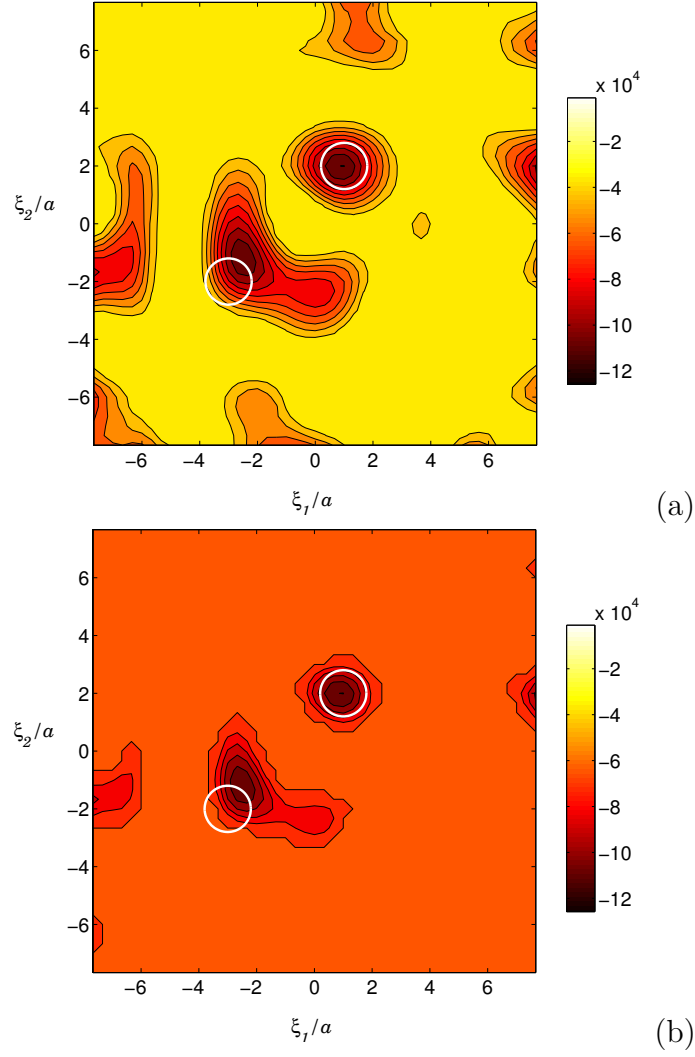


Fig. 8. Identification of two spherical hard scatterers of radius $0.8a$: distribution of thresholded topological sensitivity $\hat{\mathcal{T}}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate plane $\xi_3 = x_3^{\text{true}}$, with threshold $C = 0.25$ (a) or $C = 0.5$ (b).

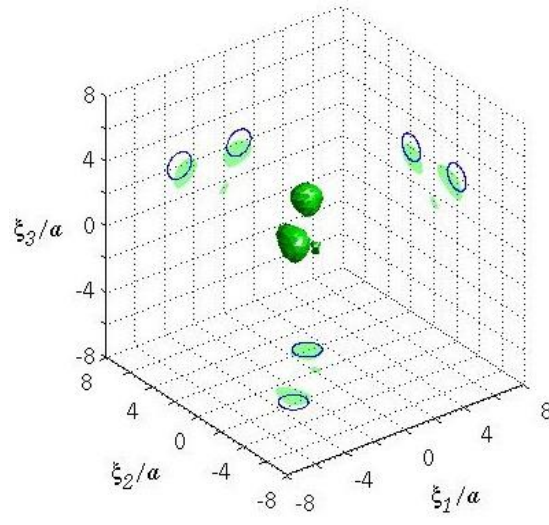


Fig. 9. Identification of two spherical hard scatterers of radius $0.8a$: iso-surface of $\mathcal{T}(\mathbf{x}_s)$ for $\mathcal{T} = 0.75\mathcal{T}_{\min}$ on the truncated search grid such that $\{|\xi_i| \leq 7a \ (i = 1, 2, 3)\}$. Values of $\mathcal{T}(\mathbf{x}_s)$ lower than the iso-value are inside the iso-surface.

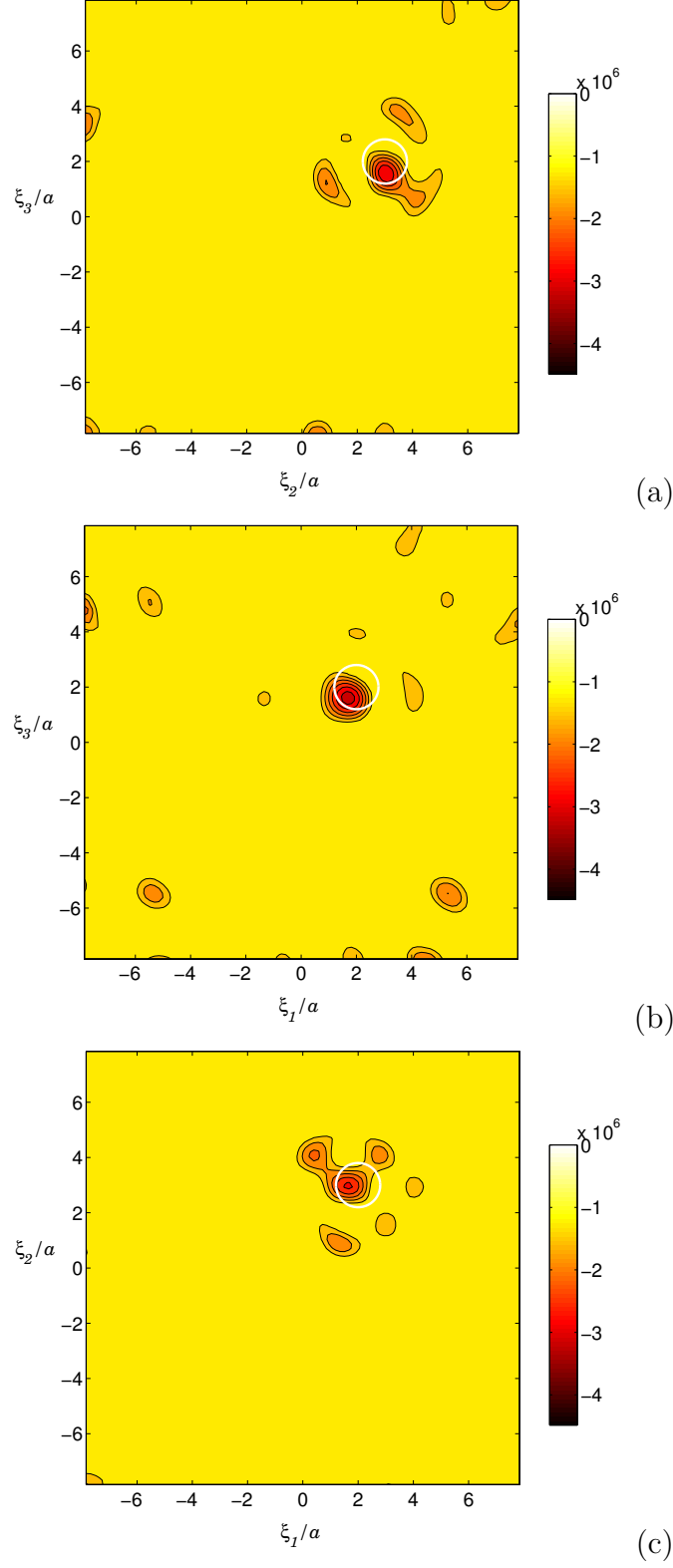


Fig. 10. Identification of spherical hard scatterer of radius $0.4a$, synthetic data with 10% noise: distribution of thresholded topological sensitivity $\hat{T}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

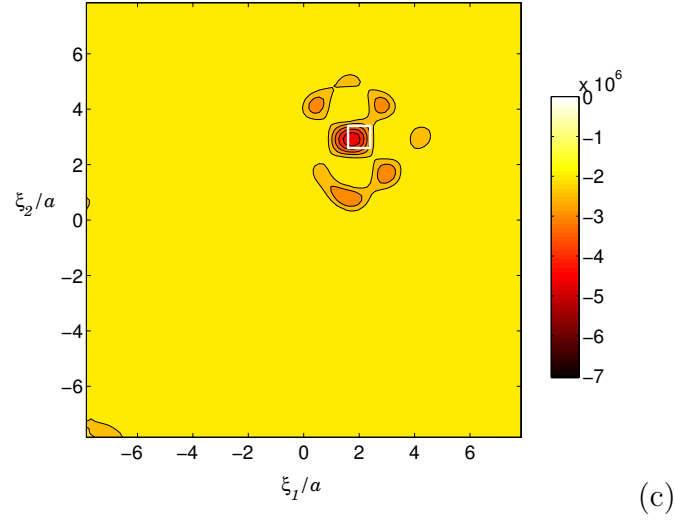
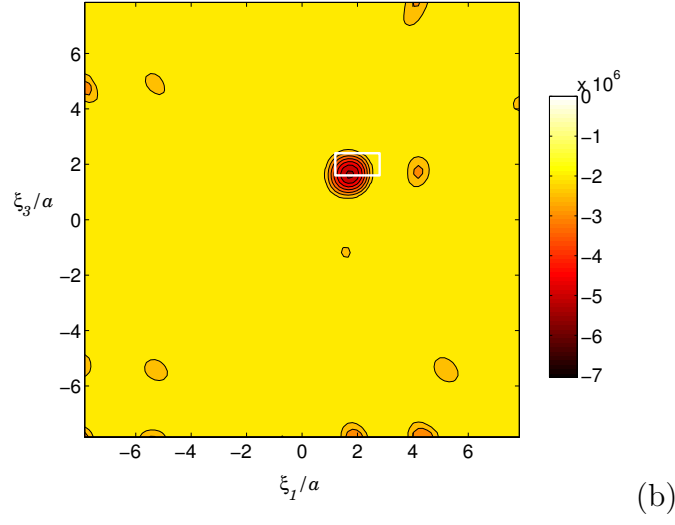
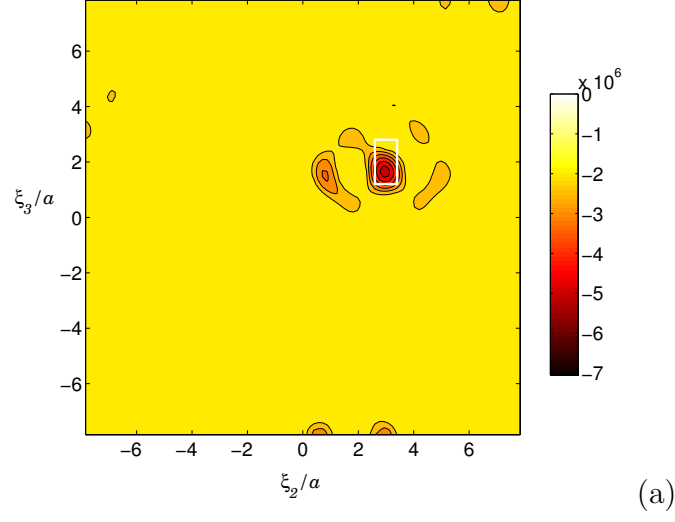


Fig. 11. Identification of box-shaped scatterer: distribution of thresholded topological sensitivity $\hat{\mathcal{T}}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

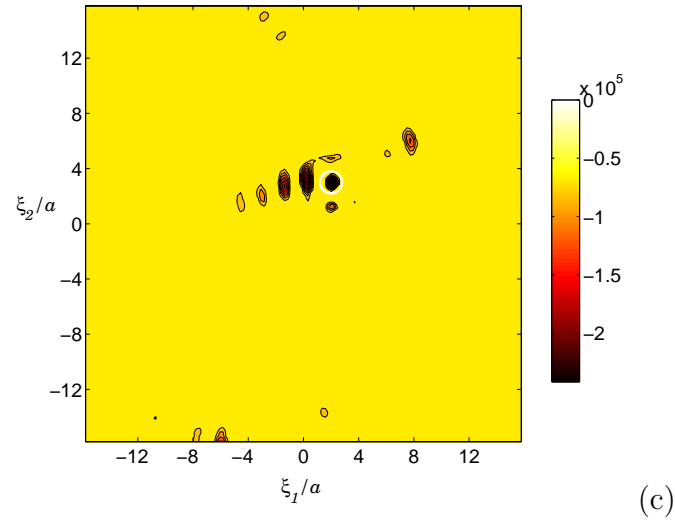
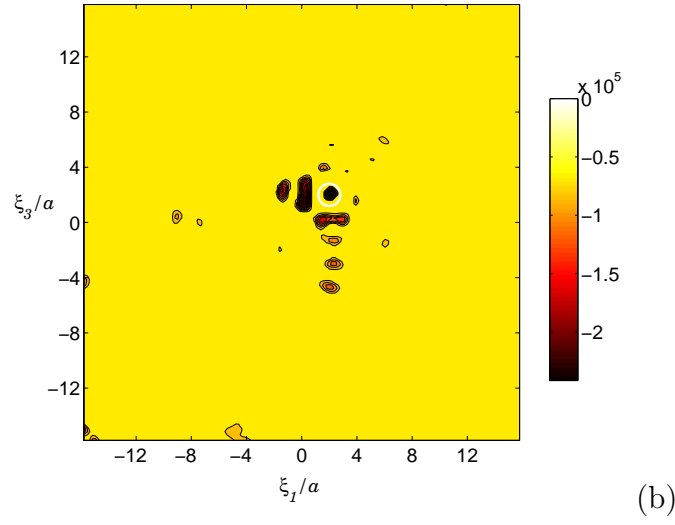
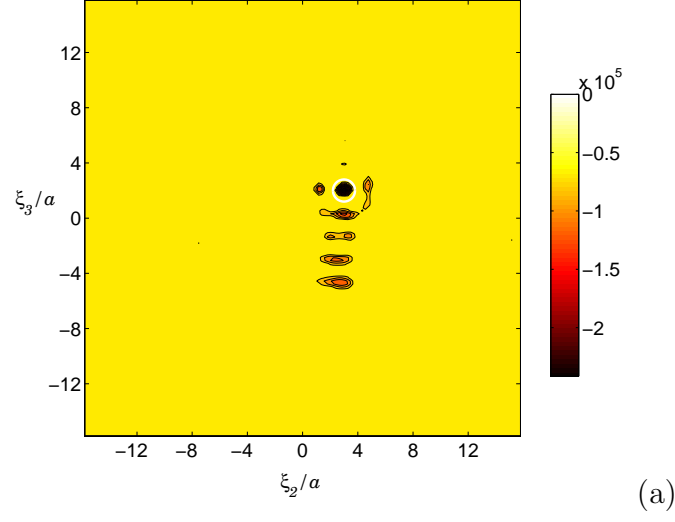


Fig. 12. Identification of spherical hard scatterer of radius $0.4a$ in large domain: distribution of thresholded topological sensitivity $\hat{\mathcal{T}}(\mathbf{x}_s)$ for sampling points \mathbf{x}_s in coordinate planes $\xi_1 = x_1^{\text{true}}$ (a), $x_2 = x_2^{\text{true}}$ (b) and $x_3 = x_3^{\text{true}}$ (c).

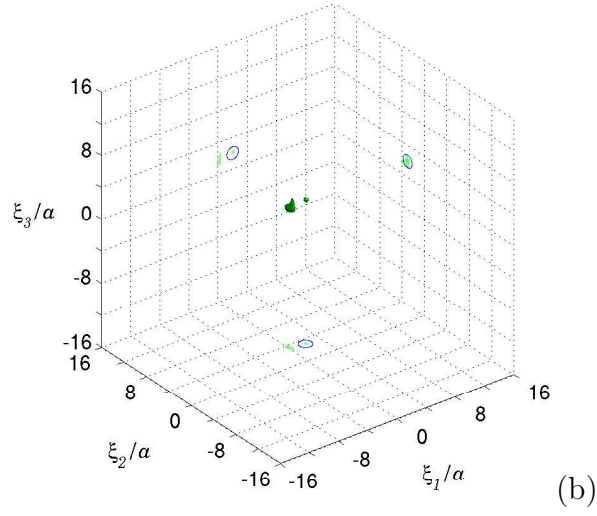
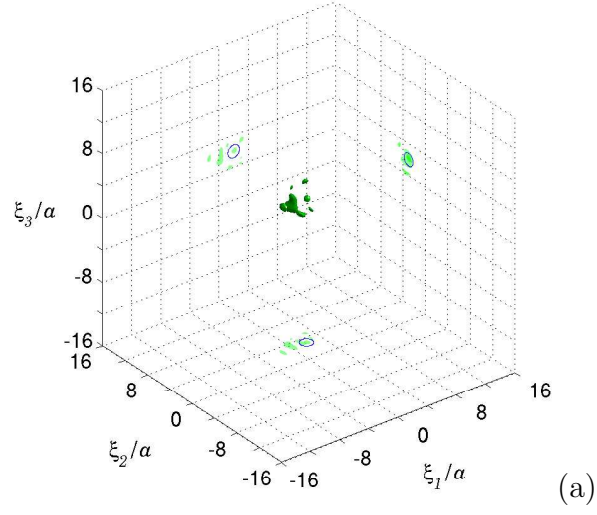
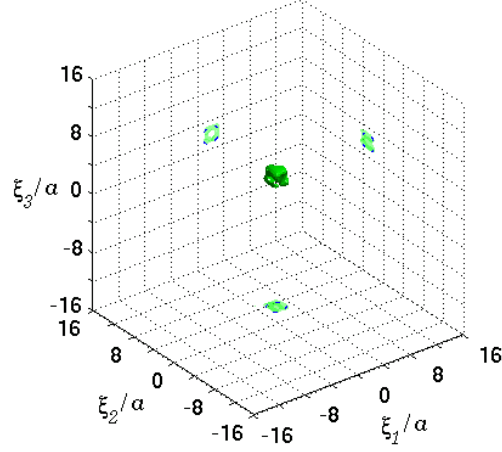
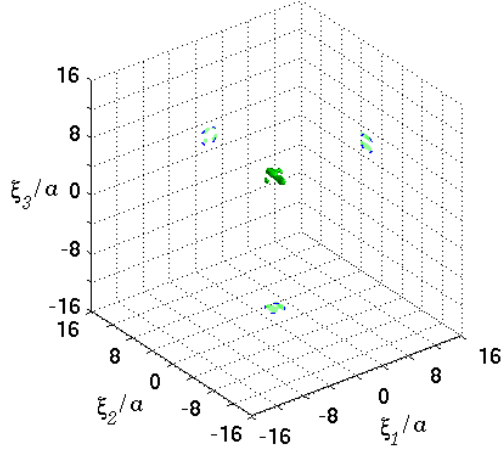


Fig. 13. Identification of spherical hard scatterer of radius $0.4a$ embedded in large domain $\Omega(16a)$: iso-surfaces of (a) $\mathcal{T}(\mathbf{x}_s)$ for $\mathcal{T} = 0.55\mathcal{T}_{\min}$ and (b) $\mathcal{T} = 0.75\mathcal{T}_{\min}$ computed on the truncated search grid such that $\{|\xi_i| \leq 14a \ (i = 1, 2, 3)\}$. Values of $\mathcal{T}(\mathbf{x}_s)$ lower than the iso-value are inside the iso-surface.



(a)



(b)

Fig. 14. Identification of spherical hard scatterer of radius $1.2a$ embedded in large domain $\Omega(16a)$: iso-surfaces of (a) $\mathcal{T}(\mathbf{x}_s)$ for $\mathcal{T} = 0.55\mathcal{T}_{\min}$ and (b) $\mathcal{T} = 0.75\mathcal{T}_{\min}$ computed on the truncated search grid such that $\{|\xi_i| \leq 14a \ (i = 1, 2, 3)\}$. Values of $\mathcal{T}(\mathbf{x}_s)$ lower than the iso-value are inside the iso-surface.